

Amendments to the Claims

This listing of claims will replace all prior versions and listings of claims in the application.

Listing of Claims:

1. (Currently Amended) An image reconstruction method from electrical capacitance tomography data comprising: (a) obtaining the measurement electrical capacitance tomography data by using a recording sensor formed by an array of electrodes located on the perimeter of an oil-pipe, well or tank corresponding to the image; (b) processing said measurement data using heuristic non-linear global optimization, specifically the Method of Simulated Annealing for the estimation of an electrical permittivity distribution image in order to obtain an image; and (c) displaying said processed image on a display device in order to visualize multiphase oil-flows through a cross section of a pipe, well or tank.

2. (Currently Amended) A method according to claim 1, wherein the measurement data are collected on the perimeter of a region such as the inside of an oil-pipeline, well or tank. electrical parameters on the perimeter of a region such as the inside of a pipeline, of the oil well or of the a tank, for the obtained image of the spatial distribution of the electrical permittivity (or dielectric constant) or of the electrical conductivity within said region, and the image reflects the spatial distribution of the materials or substances that occupy the region, such as gases or/and liquids.

3. (Currently Amended) A method according to claim 2, wherein said parameters measurements are the electrical capacitance values recorded~~measured~~ between the electrodes of a sensor, which is formed by a plurality of said electrodes placed around the perimeter of the region (pipeline, well, tank).

4. (Currently Amended) A method according to claim 1, wherein said estimated image represents image obtained is an image of the spatial distribution of the electrical permittivity in a cross sectional region of an oil-pipeline, well, or tank, which reflects the spatial distribution of the materials or substances, such as gases and/or liquids, that are

flowing through this region (pipeline, well, tank). ~~(or dielectric constant) within said interest region (pipeline, well, tank), which reflects the spatial distribution of the materials or substances, such as gases or/and liquids, that occupy this region (pipeline, well, tank).~~

5. (Currently Amended) A method according to claim 2, wherein said permittivity image is formed by a finite number of sub-regions or pixels in the oil-flow viewing, which number depends on the desired resolution permittivity image resolution.

6. (Original) A method according to claim 3, wherein said sensor is formed by a pipe made of an electrically insulating material, on whose outer wall an array of rectangular metallic electrodes is placed.

7. (Previously Presented) A method according to claim 3, wherein the sensor contains a multiphase or multicomponent flow, and images that show the distribution of the phases or components, such as gases or/and liquids, are obtained in accordance to what is said in the background of the invention.

8. (Cancelled)

9. (Currently Amended) A method according to claim [[8]]1, wherein a cost function associated to the energy of the system is iteratively minimized with respect to ϵ (the vector of the permittivity values in each pixel of the image), said cost function being of the form:

$$E_k = L_{2(k)} = \frac{\sum_{i=1}^m [c_i^{meas} - c_i^{calc}(\epsilon_k)]^2}{\sum_{i=1}^m [c_i^{meas}]^2} \quad (i = 1, \dots, m)$$

where c_i^{meas} are the m measured mutual capacitances and $c_i^{calc}(\epsilon_k)$ are the ones calculated by solving the forward problem for a given permittivity distribution ϵ_k .

10. (Original) A method according to claim 9, wherein the Metropolis criterion is used to do the minimization.

11. (Original) A method according to claim 10, wherein in the minimization process it is used, as the initial guess for the permittivity distribution, ϵ , both a homogeneous distribution and the distribution that results from applying the linear back-projection (LBP) method to the measurement data.

12. (Original) A method according to claim 9, wherein the computation of the calculated capacitances $c_i^{calc}(\epsilon_k)$, known as the forward problem, is carried out by means of the finite-volume method.

13. (Original) A method according to claim 12, wherein for the solution of the system of equations that results from solving the forward problem $c_i^{calc}(\epsilon_k)$, iterative methods that rapidly converge are used, by employing as initial estimation of the electrostatic potential the result of the forward problem solution obtained in the previous iteration of the inverse problem.

14. (Currently Amended) A method according to claim 1, wherein the method of genetic algorithms is used as an alternative electrical permittivity image estimation~~the optimization~~ method.

15. (Original) A method according to claim 14, wherein, starting from an initial population of Q permittivity models $\epsilon_{k(o)}$ ($k = 1, \dots, Q$), evolutionary mechanisms such as selection, crossover and mutation are applied in order to obtain new populations.

16. (Original) A method according to claim 15, wherein individuals ϵ_k are characterized by having a small cost (or misfit) function, which is given by

$$E_k = L_{2(k)} = \frac{\sum_{i=1}^m [c_i^{meas} - c_i^{calc}(\epsilon_k)]^2}{\sum_{i=1}^m [c_i^{meas}]^2} \quad (i = 1, \dots, m)$$

where c_i^{meas} are the m measured mutual capacitances and $c_i^{calc}(\epsilon_k)$ are the calculated ones by solving the forward problem for a given model or permittivity distribution ϵ_k .

17. (Previously Presented) A method according to claim 15, wherein the accumulated probability of selection for a particular model ϵ_k is given by

$$P(\epsilon_k) = P(\epsilon_{k-1}) + \frac{E_{max} - E(\epsilon_k)}{Q (E_{max} - E_{avr})}$$

where E_{max} , and E_{avr} are maximum and average cost functions of the generation, respectively, and Q is the number of individuals in the population.

18. (Original) A method according to claim 17, wherein a biased roulette procedure can be used to decide which models are selected on each iteration of the method.

19. (Original) A method according to claim 18, wherein crossover and mutation of models are carried out randomly according to the probabilities of crossover and mutation, P_c y P_m .

20. (Original) A method according to claim 19, wherein the probability of mutation P_m is determined using the average variation coefficient γ , given by

$$\gamma = \frac{1}{p} \sum_{i=1}^p \left(\frac{\sigma_i}{\bar{\epsilon}_i} \right)$$

where p is the number of parameters, $\bar{\epsilon}_i$ is the average of the i -th parameter, and σ_i is the standard deviation.

21. (Previously Presented) A method according to claim 19, wherein P_m is defined as a function of γ , that is:

$$P_m = \begin{cases} P_{ini} & \text{para } \gamma > 0.1 \\ 0.1 & \text{para } 0.02 < \gamma < 0.1 \\ 0.2 & \text{para } \gamma < 0.02 \end{cases}$$

where P_{ini} is the initial probability of mutation.

22. (Previously Presented) A method according to claim 15, wherein the calculation of $c_i^{calc}(\epsilon_k)$, known as the forward problem, is performed by means of the finite-volume method.